

Problem 1

We define the random variables S ("Scanner") and T ("Terrorist") as follows

$$\begin{aligned} S(\text{'person is predicted as terrorist'}) &= 1 \\ S(\text{'person is predicted as not terrorist'}) &= 0 \end{aligned}$$

$$\begin{aligned} T(\text{'person is an actual terrorist'}) &= 1 \\ T(\text{'person is not an actual terrorist'}) &= 0 \end{aligned}$$

By problem statement we have

$$\begin{aligned} P(T = 1) &= 0.01 \Rightarrow P(T = 0) = 0.99 \\ P(S = 1|T = 1) &= 0.95 \Rightarrow P(S = 0|T = 1) = 0.05 \\ P(S = 0|T = 0) &= 0.95 \Rightarrow P(S = 1|T = 0) = 0.05 \end{aligned}$$

Goal: Find $P(T = 1|S = 1)$.

Solution.

$$\begin{aligned} P(T = 1|S = 1) &= \frac{P(S = 1|T = 1) \cdot P(T = 1)}{P(S = 1)} && \text{(by Bayes' Theorem)} \\ &= \frac{P(S = 1|T = 1) \cdot P(T = 1)}{P(S = 1|T = 1) \cdot P(T = 1) + P(S = 1|T = 0) \cdot P(T = 0)} \\ &&& \text{(by Law of Total Probability)} \\ &= \frac{0.95 \cdot 0.01}{0.95 \cdot 0.01 + 0.05 \cdot 0.99} \\ &= 0.16 \end{aligned}$$

Problem 2

Let $\Omega = \{w, r\}$. Let's define all possible events of putting two balls in the basket.

$$P("rr") = P("wr") = P("rw") = P("ww") = \frac{1}{4}$$

Let's define also $\Omega' = \{W, R\}$ and let "RRR" be the event of drawing three red balls from the basket.

Goal: Find $P("rr"|"RRR")$.

Solution.

$$\begin{aligned} P("rr"|"RRR") &= \frac{P("RRR"|"rr") \cdot P("rr")}{P("RRR")} && \text{(by Bayes' Theorem)} \\ &= \frac{1 \cdot 1/4}{1 \cdot 1/4 + 1/8 \cdot 1/4 + 1/8 \cdot 1/4 + 0 \cdot 1/4} && \text{(by Law of Total Probability)} \\ &= 0.8 \end{aligned}$$

Problem 3

$$X(X=x) = \begin{cases} 1, & \text{if } x = \text{"H"} \\ 0, & \text{if } x = \text{"T"} \end{cases}$$

$Y(Y = y_i) = i$, if $y_i = \text{"Coin X is tossed } i \text{ times, where the first } i-1 \text{ are T and the } i\text{-th is H"}$.

Solution.

$$\mathbb{E}_Y[X = 1] = 1 \quad (\text{by problem statement})$$

$$\begin{aligned} \mathbb{E}_Y[X = 0] &= \sum_{i=0}^{\infty} \cdot \left(\frac{1}{2}\right)^i \cdot \frac{1}{2} && (\text{by expectation definition}) \\ &= \frac{1}{2} \cdot \sum_{i=0}^{\infty} \cdot \left(\frac{1}{2}\right)^i \\ &= \frac{1}{2} \cdot \frac{1}{1 - \frac{1}{2}} && (\text{by geometric series definition}) \\ &= 1 \end{aligned}$$

Problem 4

Goal: Compute $\mathbb{E}[X]$, $Var[X]$.

Solution.

$$\begin{aligned} \mathbb{E}[X] &= \int_{-\infty}^{\infty} x \cdot p(x) \, dx && (\text{by definition expectation of continuous random variable}) \\ &= \int_a^b x \cdot p(x) \, dx && (\text{by definition } p(x) \text{ is zero outside these boundaries}) \\ &= \frac{1}{b-a} \int_a^b x \, dx \\ &= \frac{1}{b-a} \left[\frac{x^2}{2} \right]_a^b \\ &= \frac{1}{b-a} \cdot \frac{b^2 - a^2}{2} \\ &= \frac{a+b}{2} \end{aligned}$$

$$\begin{aligned}
\text{Var}[X] &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 && \text{(by definition variance of continuous random variable)} \\
&= \mathbb{E}[X^2] - \left(\frac{a+b}{2}\right)^2 \\
&= \int_{-\infty}^{\infty} x^2 \cdot p(x) \, dx - \left(\frac{a+b}{2}\right)^2 \\
&= \int_a^b x^2 \cdot p(x) \, dx - \left(\frac{a+b}{2}\right)^2 && \text{(by definition } p(x) \text{ is zero outside these boundaries)} \\
&= \frac{1}{b-a} \int_a^b x^2 \, dx - \left(\frac{a+b}{2}\right)^2 \\
&= \frac{1}{b-a} \left[\frac{x^3}{3} \right]_a^b - \left(\frac{a+b}{2}\right)^2 \\
&= \frac{1}{b-a} \cdot \frac{b^3 - a^3}{3} - \left(\frac{a+b}{2}\right)^2 \\
&= \frac{b^2 + ab + a^2}{3} - \left(\frac{a+b}{2}\right)^2 && \text{(by difference of cubes)} \\
&= \frac{b^2 - 2ab + a^2}{12} \\
&= \frac{(a-b)^2}{12}
\end{aligned}$$

Problem 5

Goal: Prove

$$\begin{aligned}
\mathbb{E}[X] &= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X]] \\
\text{Var}[X] &= \mathbb{E}_Y[\text{Var}_{X|Y}[X]] + \text{Var}_Y[\mathbb{E}_{X|Y}[X]]
\end{aligned}$$

Solution. For simplicity, we prove for the discrete case however going to continuous case requires only the use of integrals instead of sums.

$$\begin{aligned}
\mathbb{E}_Y[\mathbb{E}_{X|Y}[X]] &= \sum_y p(y) \cdot \mathbb{E}_{X|Y}[X] \\
&= \sum_y p(y) \cdot \left(\sum_x x \cdot p(x|y) \right) \\
&= \sum_y \sum_x x \cdot p(x|y) \cdot p(y) \\
&= \sum_y \sum_x x \cdot p(x, y) \\
&= \sum_x \sum_y x \cdot p(x, y) \\
&= \sum_x x \sum_y p(x, y) \\
&= \sum_x x \cdot p(x) \\
&= \mathbb{E}[X]
\end{aligned}$$

$$\begin{aligned}
\mathbb{E}_Y[\text{Var}_{X|Y}[X]] + \text{Var}_Y[\mathbb{E}_{X|Y}[X]] &= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2] - (\mathbb{E}_{X|Y}[X])^2] + \mathbb{E}_Y[(\mathbb{E}_{X|Y}[X])^2] - (\mathbb{E}_Y[\mathbb{E}_{X|Y}[X]])^2 \\
&= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2]] - \mathbb{E}_Y[(\mathbb{E}_{X|Y}[X])^2] + \mathbb{E}_Y[(\mathbb{E}_{X|Y}[X])^2] - (\mathbb{E}_Y[\mathbb{E}_{X|Y}[X]])^2 \\
&= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2]] - (\mathbb{E}_Y[\mathbb{E}_{X|Y}[X]])^2 \\
&= \mathbb{E}[X^2] - (\mathbb{E}[X])^2 \quad (\text{by first proof}) \\
&= \text{Var}[X]
\end{aligned}$$

Problem 6

Goal: Prove $p(|\frac{1}{n} \sum_{i=1}^n X_i - \mathbb{E}[X_i]| > \epsilon) \rightarrow 0$.

Solution.

Let $X = \frac{1}{n} \sum_{i=1}^n X_i$.

$$\begin{aligned}
p\left(\left|\frac{1}{n} \sum_{i=1}^n X_i - \mathbb{E}[X_i]\right| > \epsilon\right) &= p\left(\left|X - \frac{1}{n} \sum_{i=1}^n \mathbb{E}[X_i]\right| > \epsilon\right) && (\text{linearity of expectation}) \\
&= p\left(\left|X - \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n X_i\right]\right| > \epsilon\right) && (\text{by substitution}) \\
&= p(|X - \mathbb{E}[X]| > \epsilon) && (\text{by Chebyshev inequality}) \\
&\leq \frac{\text{Var}(X)}{\epsilon^2} && (\text{by linearity of variance under i.i.d assumption}) \\
&= \frac{\frac{1}{n^2} \sum_{i=1}^n \text{Var}(X_i)}{\epsilon^2} && (X_i \text{ have same variance } \sigma^2) \\
&= \frac{\sigma^2}{\epsilon^2 n} \rightarrow 0 && (n \rightarrow \infty)
\end{aligned}$$
